

Decision Trees

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5/3/2020

Decision Trees

Import Data

```
library(rpart)
library(rpart.plot)
library(ggplot2)
data <- read.csv("./Project 02/nwCrow_bloodParasites_alaska_smith_2007_2008/nwCrow_sampling_alaska_smith_2007_2008.csv")
data_2 <- read.csv("./Project 02/nwCrow_bloodParasites_alaska_smith_2007_2008/nwCrow_bloodParasites_alaska_smith_2007_2008.csv")
de <- merge(data, data_2, by=0, all=TRUE)
head(de)
```

[illegible]

One Hot Encoding

```
for(unique_value in unique(de$LOC)){

de[paste("LOC", unique_value, sep = ".")] <- ifelse(de$LOC == unique_value, 1, 0)

}

head(de)
```

```
## Row.names Field.ID DATE LOC LAT LONG SEX AGE AKD TARSUS WING MASS
## 1 1 75001 3/20/2007 SEWA 60.11 -149.44 1 1 1 55.8 283 448
## 2 10 75010 3/22/2007 KENA 60.55 -151.23 2 1 0 48.6 271 390
## 3 100 75100 3/12/2008 VALD 61.12 -146.35 2 1 0 44.6 264 317
## 4 101 86701 3/12/2008 VALD 61.12 -146.35 2 1 0 47.1 269 343
## 5 102 86702 3/12/2008 VALD 61.12 -146.35 2 2 0 52.2 291 415
## 6 103 86703 3/12/2008 VALD 61.12 -146.35 1 2 0 47.0 266 325
## Extraction.. LEUC1 LEUC2 HAEM1 HAEM2 PLAS1 PLAS2 Leuc_GenBank_Accession
## 1 NOCR001 0 0 0 0 0 0
## 2 NOCR010 0 0 0 0 0 0
## 3 NOCR100 1 1 0 0 0 0 MG765394
## 4 NOCR101 0 0 0 0 0 0
## 5 NOCR102 1 0 0 0 0 0 MG765394
## 6 NOCR103 1 1 0 0 0 0 MG765394
## Haem_GenBank_Accession Plas_GenBank_Accession LOC.SEWA LOC.KENA LOC.VALD
## 1 1 0 0
## 2 0 1 0
## 3 0 0 1
## 4 0 0 1
## 5 0 0 1
## 6 0 0 1
## LOC.HAIN LOC.JUNE LOC.HOME
## 1 0 0 0
## 2 0 0 0
## 3 0 0 0
## 4 0 0 0
## 5 0 0 0
## 6 0 0 0
```

Filter Columns

```
de <- de[,c(7,8,9,10,11,12,14,16,18,23,24,25,26,27,28)]
head(de)
```

```
## SEX AGE AKD TARSUS WING MASS LEUC1 HAEM1 PLAS1 LOC.SEWA LOC.KENA LOC.VALD
## 1 1 1 1 55.8 283 448 0 0 0 1 0 0
## 2 2 1 0 48.6 271 390 0 0 0 0 1 0
## 3 2 1 0 44.6 264 317 1 0 0 0 0 1
## 4 2 1 0 47.1 269 343 0 0 0 0 0 1
## 5 2 2 0 52.2 291 415 1 0 0 0 0 1
## 6 1 2 0 47.0 266 325 1 0 0 0 0 1
## LOC.HAIN LOC.JUNE LOC.HOME
## 1 0 0 0
```

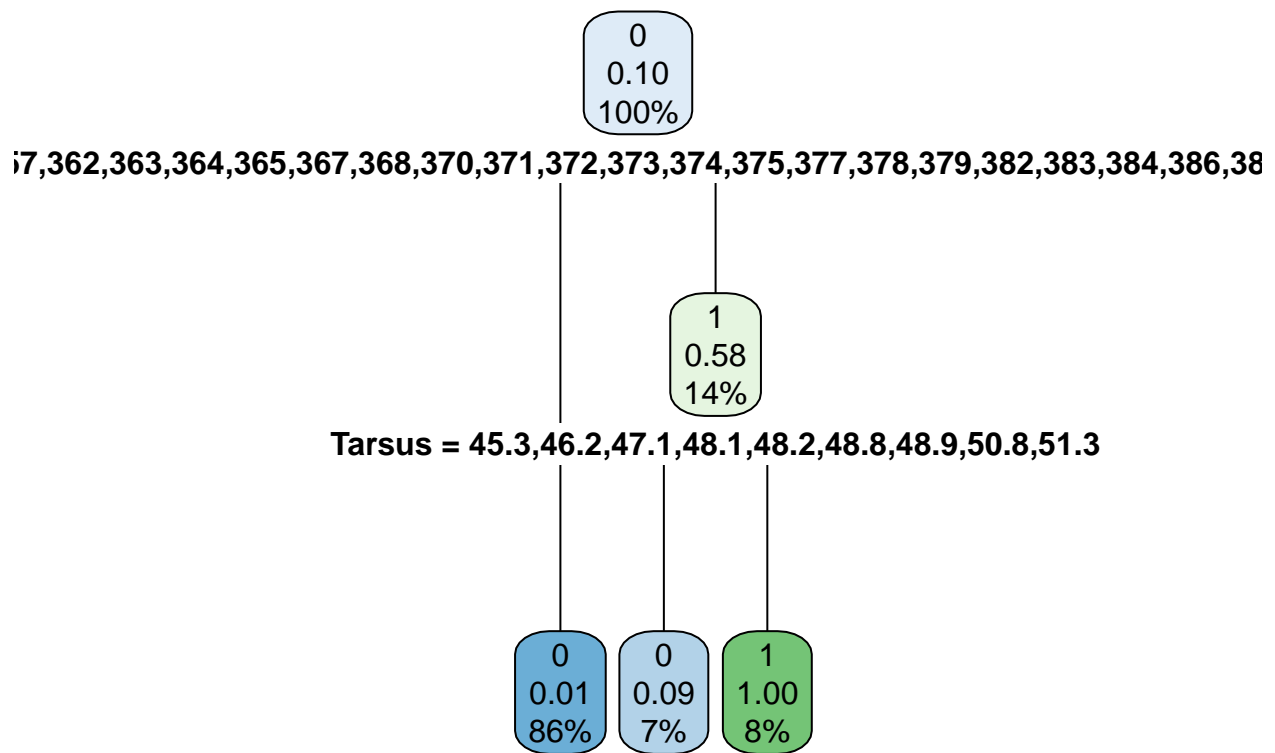
```
## 2      0      0      0
## 3      0      0      0
## 4      0      0      0
## 5      0      0      0
## 6      0      0      0
```

Factoring

```
de$SEX <- as.factor(de$SEX)
de$AGE <- as.factor(de$AGE)
de$AKD <- as.factor(de$AKD)
de$TARSUS <- as.factor(de$TARSUS)
de$WING <- as.factor(de$WING)
de$MASS <- as.factor(de$MASS)
de$LEUC1 <- as.factor(de$LEUC1)
de$HAEM1 <- as.factor(de$HAEM1)
de$PLAS1 <- as.factor(de$PLAS1)
de$LOC.SEWA <- as.factor(de$LOC.SEWA)
de$LOC.KENA <- as.factor(de$LOC.KENA)
de$LOC.VALD <- as.factor(de$LOC.VALD)
de$LOC.HAIN <- as.factor(de$LOC.HAIN)
de$LOC.JUNE <- as.factor(de$LOC.JUNE)
de$LOC.HAIN <- as.factor(de$LOC.HAIN)
```

AKD Decision Tree

```
names(de) <- c("Sex", "Age", "AKD", "Tarsus", "Wing", "Mass", "LEUC1", "HAEM1", "PLAS1", "SEWA", "KENA", "VALD", "H")
ran <- sample(1:nrow(de), 0.9 * nrow(de))
data_train <- de[ran,]
data_test <- de[-ran,]
dtm <- rpart(AKD~., data_train, method="class")
rpart.plot(dtm, compress=TRUE, uniform=TRUE)
```



```

p <- predict(dtm, data_test, type="class")
confMat <- table(data_test$AKD,p)
accuracy <- sum(diag(confMat))/sum(confMat)
return (accuracy*100)

```

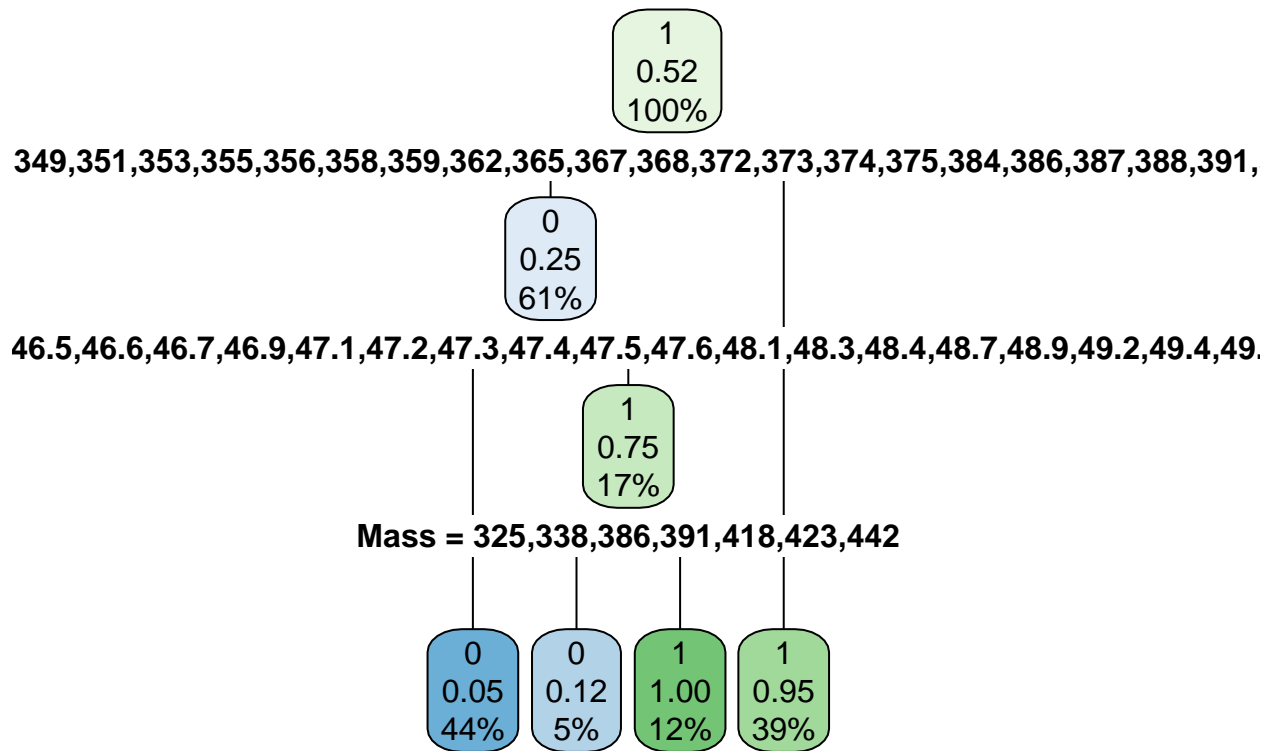
```
## [1] 89.47368
```

LEUC1 Decision Tree

```

names(de) <- c("Sex", "Age", "AKD", "Tarsus", "Wing", "Mass", "LEUC1", "HAEM1", "PLAS1", "SEWA", "KENA", "VALD", "H")
ran <- sample(1:nrow(de), 0.9 * nrow(de))
data_train <- de[ran,]
data_test <- de[-ran,]
dtm <- rpart(LEUC1~., data_train, method="class")
rpart.plot(dtm, compress=TRUE, uniform=TRUE)

```



```

p <- predict(dtm, data_test, type="class")
confMat <- table(data_test$LEUC1,p)
accuracy <- sum(diag(confMat))/sum(confMat)
return (accuracy*100)

```

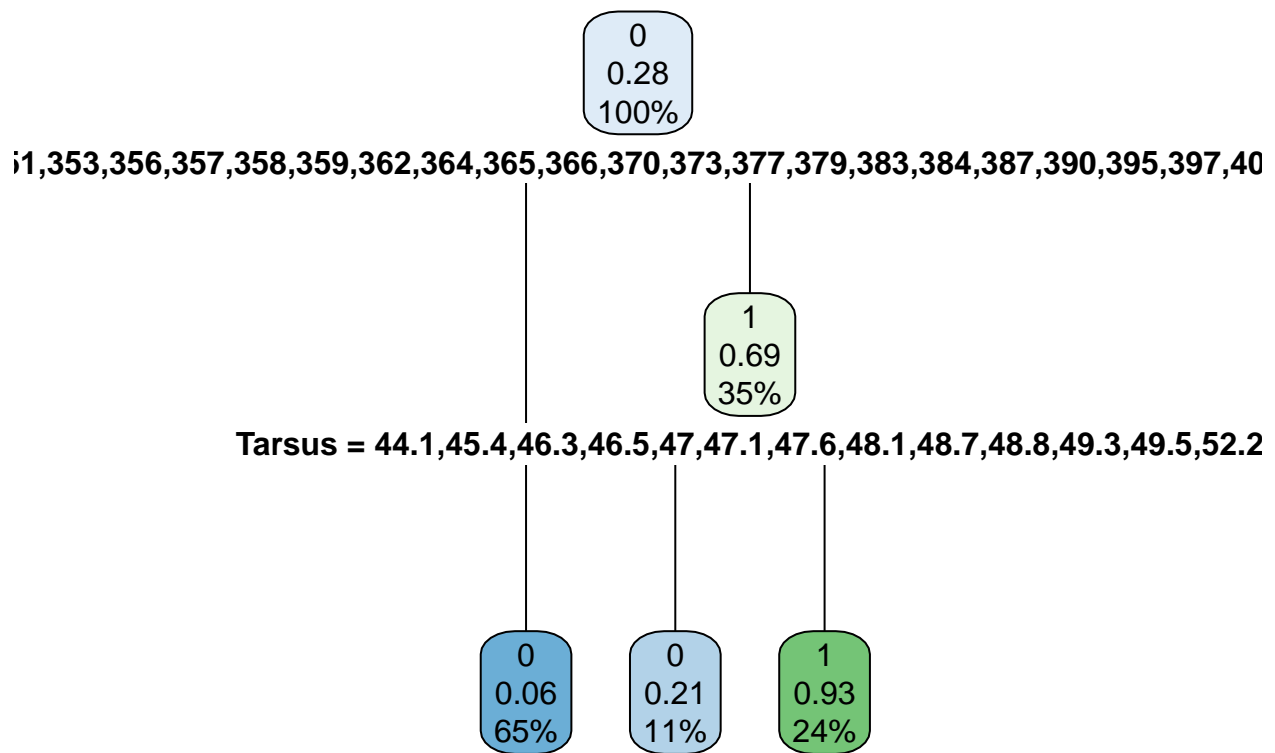
```
## [1] 42.10526
```

HAEM1 Decision Tree

```

names(de) <- c("Sex","Age","AKD","Tarsus","Wing","Mass","LEUC1","HAEM1","PLAS1","SEWA","KENA","VALD","H
ran <- sample(1:nrow(de), 0.9 * nrow(de))
data_train <- de[ran,]
data_test <- de[-ran,]
dtm <- rpart(HAEM1~., data_train, method="class")
rpart.plot(dtm, compress=TRUE, uniform=TRUE)

```

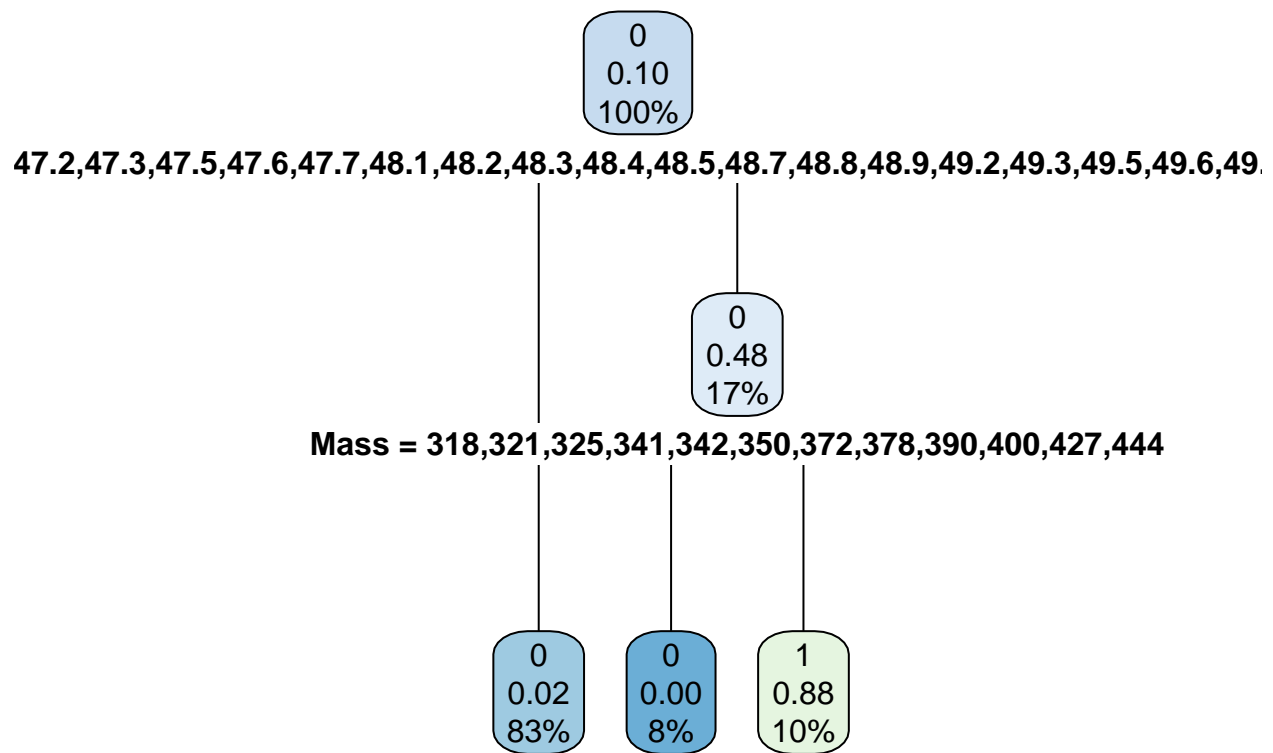


```
p <- predict(dtm, data_test, type="class")
confMat <- table(data_test$HAEM1,p)
accuracy <- sum(diag(confMat))/sum(confMat)
return (accuracy*100)
```

```
## [1] 47.36842
```

PLAS1 Decision Tree

```
names(de) <- c("Sex", "Age", "AKD", "Tarsus", "Wing", "Mass", "LEUC1", "HAEM1", "PLAS1", "SEWA", "KENA", "VALD", "H")
ran <- sample(1:nrow(de), 0.9 * nrow(de))
data_train <- de[ran,]
data_test <- de[-ran,]
dtm <- rpart(PLAS1~., data_train, method="class")
rpart.plot(dtm, compress=TRUE, uniform=TRUE)
```



```

p <- predict(dtm, data_test, type="class")
confMat <- table(data_test$PLAS1,p)
accuracy <- sum(diag(confMat))/sum(confMat)
return (accuracy*100)

```

```
## [1] 84.21053
```